

Machine Learning for the Identification of Geographic Hotspots to Deploy HIV Testing for Prevention, Early Case Finding and Targeted Post-Test Services

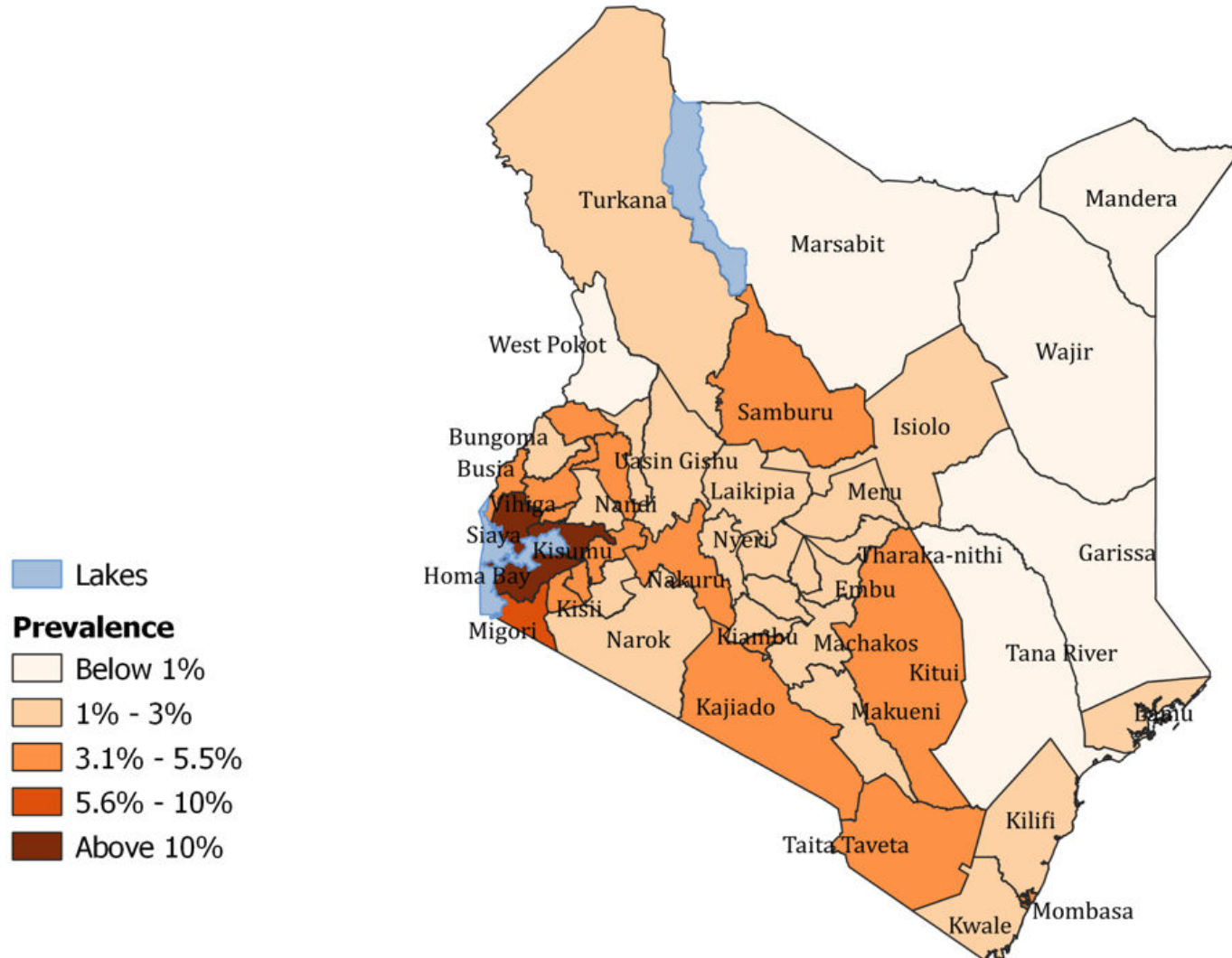
Digital Platforms to Enhance HIV Testing and Linkage Coverage

Jonathan Mwangi, CDC Kenya



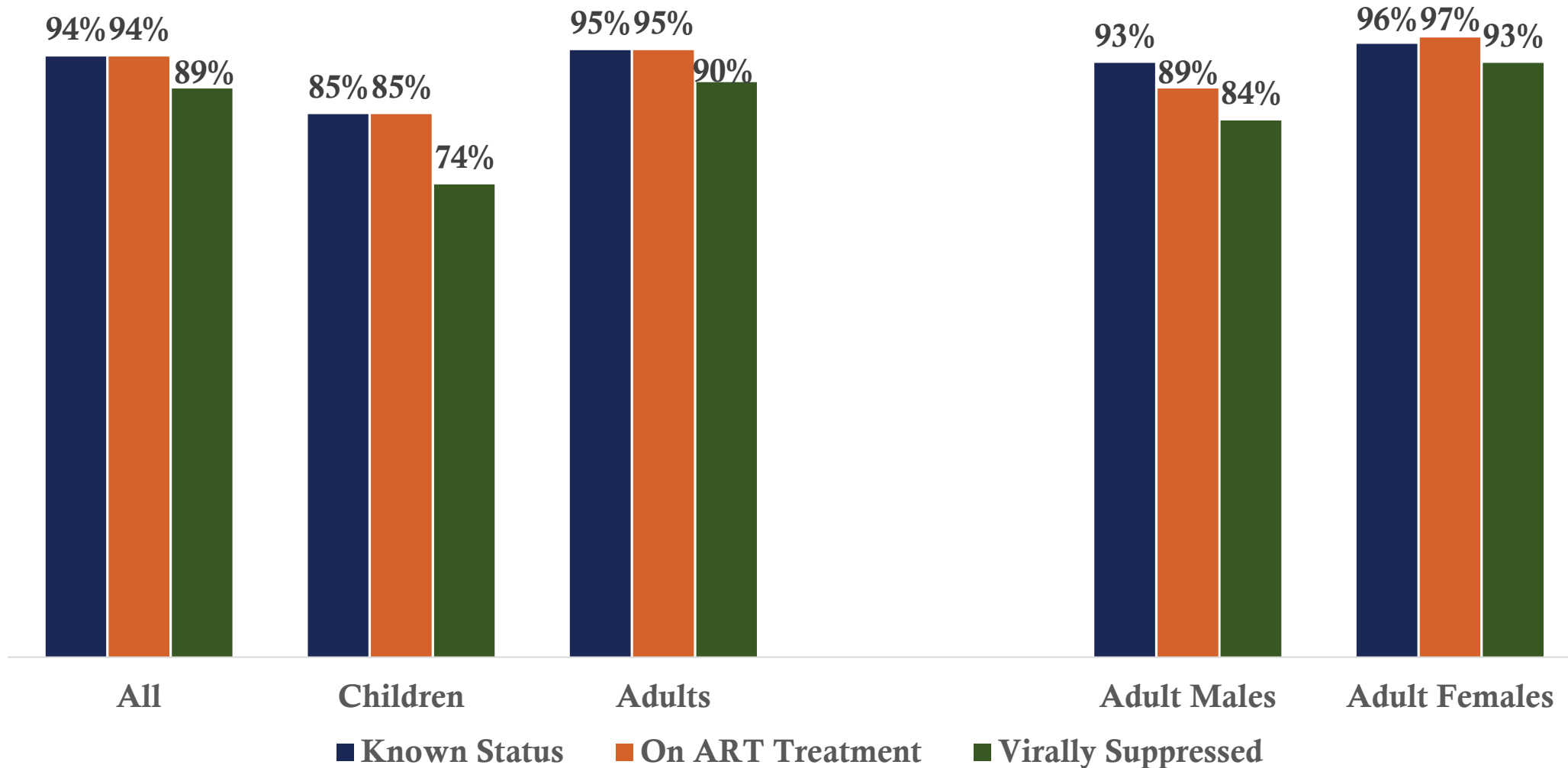
CQUIN dHTS Meeting | July 9 - 12, 2024 – Durban, South Africa

Background: Kenya Epi Profile

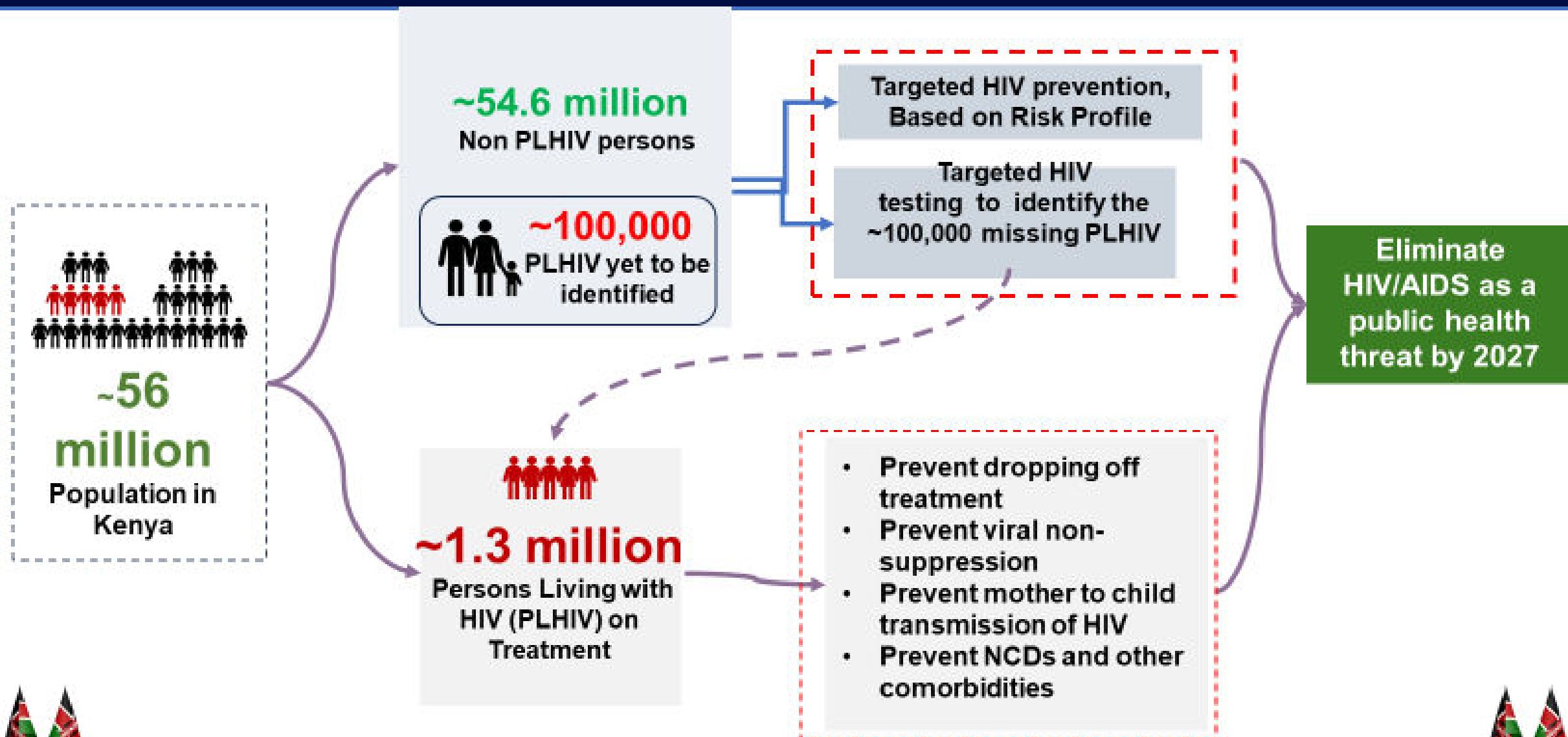


- HIV Infections-**22,154**
- HIV Prevalence- **3.7%**
- Incidence- **0.059%**
- Mortality- **18,473**
- MTCT Rate- **8.6%**
- PLHIV- **1,377,784**

Background: 95-95-95 Progress

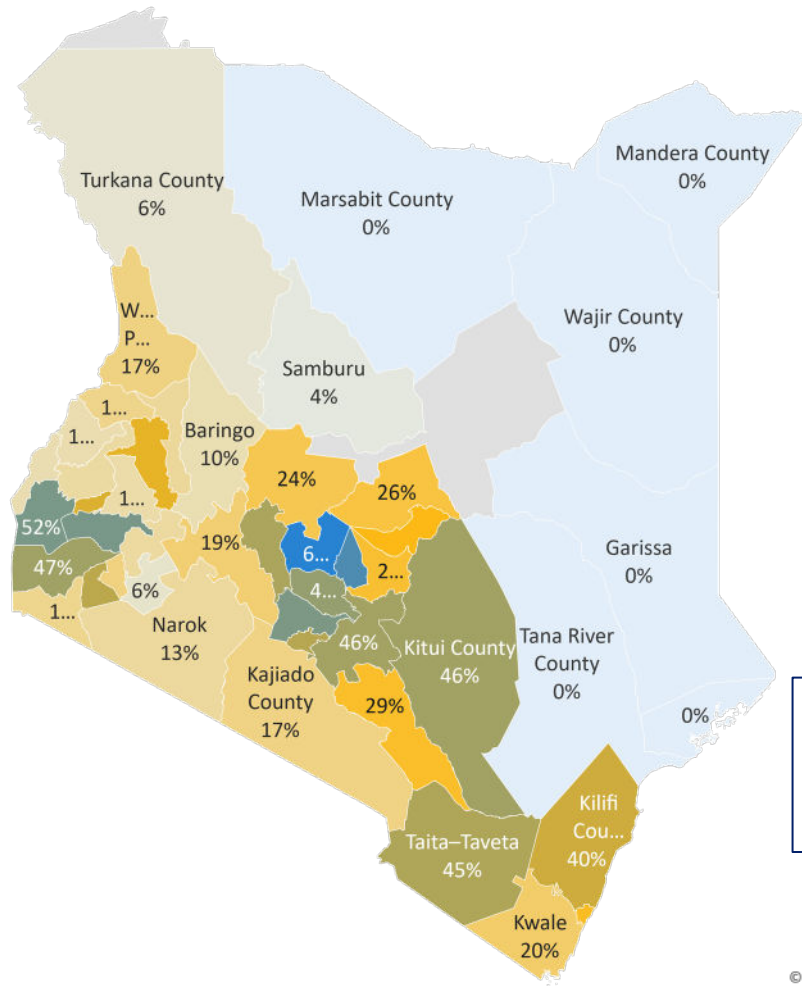


To Achieve and Sustain Epidemic Control

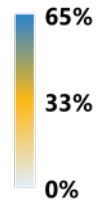


Data will be Key: % of HTS Data Captured Electronically and transmitted to the National Data Warehouse (NDW)

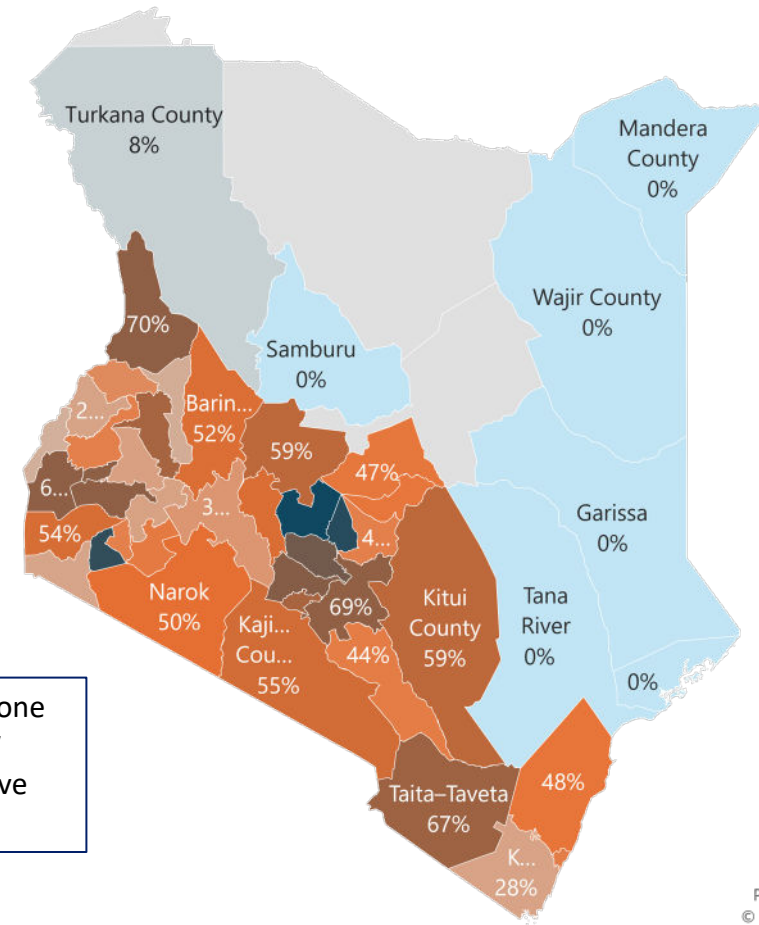
HTS_TST Coverage April 2024



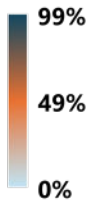
%NDW Coverage



HTS_POS Coverage April 2024



%NDW Coverage

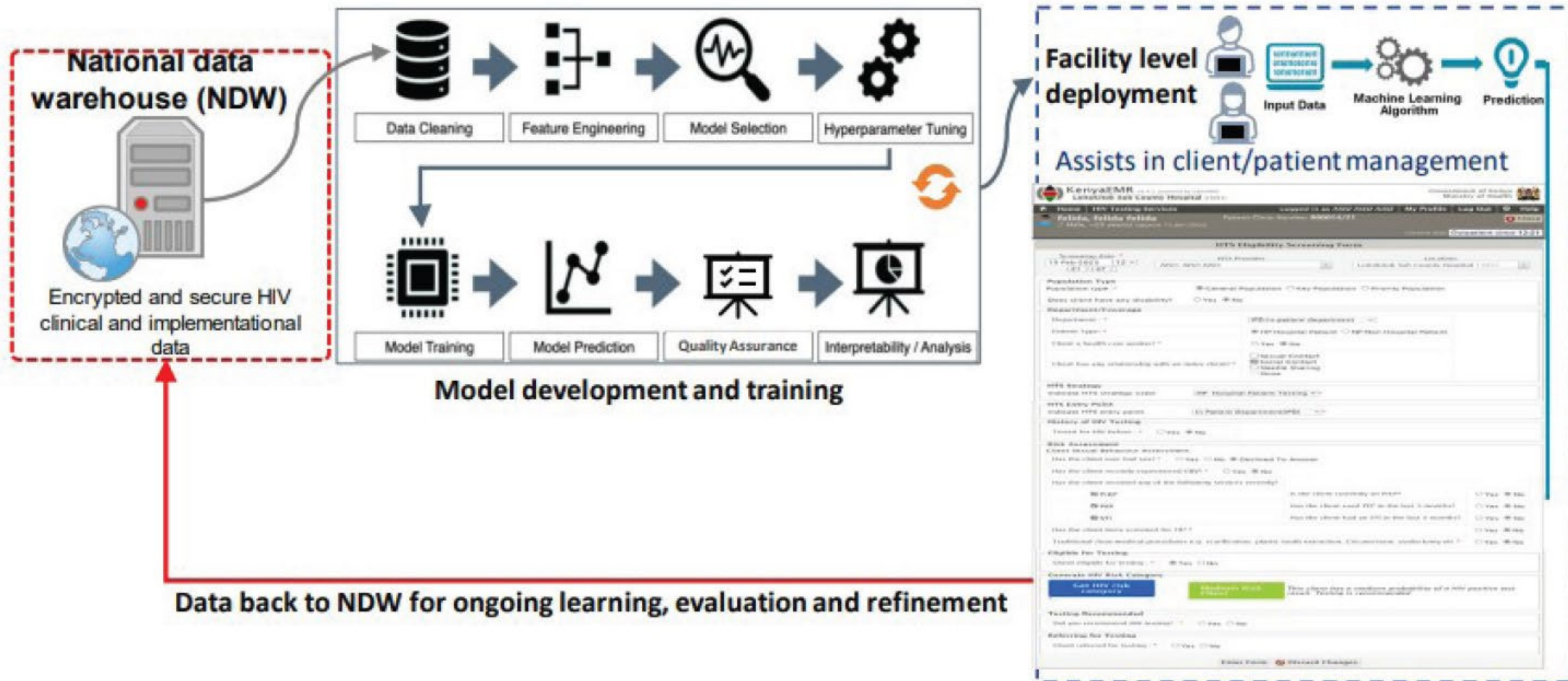


- **32%** (197,138/609,031) of tests done country wide are available in NDW
- **53%** (3,658/6,924) of tested positive countrywide are available in NDW

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Building on eHTS Platform: Machine Learning Solutions



Machine Learning for HTS

The screenshot shows the KenyaEMR interface for the HTS Eligibility Screening Form. The patient is identified as 'felida, felida felida', a male, approximately 20 years old, with a patient clinic number of 000014/21. The form is for a screening date of 15-Feb-2023 at Loitokitok Sub County Hospital 15051, with AISO, AISO AISO as the provider. The form includes sections for Population Type (General Population selected), Department/Coverage (IPD: In-patient department), HTS Strategy (HP: Hospital Patient Testing), HTS Entry Point (In Patient Department (IPD)), History of HIV Testing (Not tested before), Risk Assessment (Client Sexual Behaviour Assessment), Eligible for Testing (Yes selected), and Generate HIV Risk Category (Medium Risk Client). A note states: 'This client has a medium probability of a HIV positive test result. Testing is recommended.' The form also includes buttons for 'Enter Form' and 'Discard Changes'.

What is it?

- A HIV testing decision alert system that relies on a machine learning algorithm to classify risk of testing HIV positive.
- The algorithm uses both Sociodemographic and Behavioral variables.

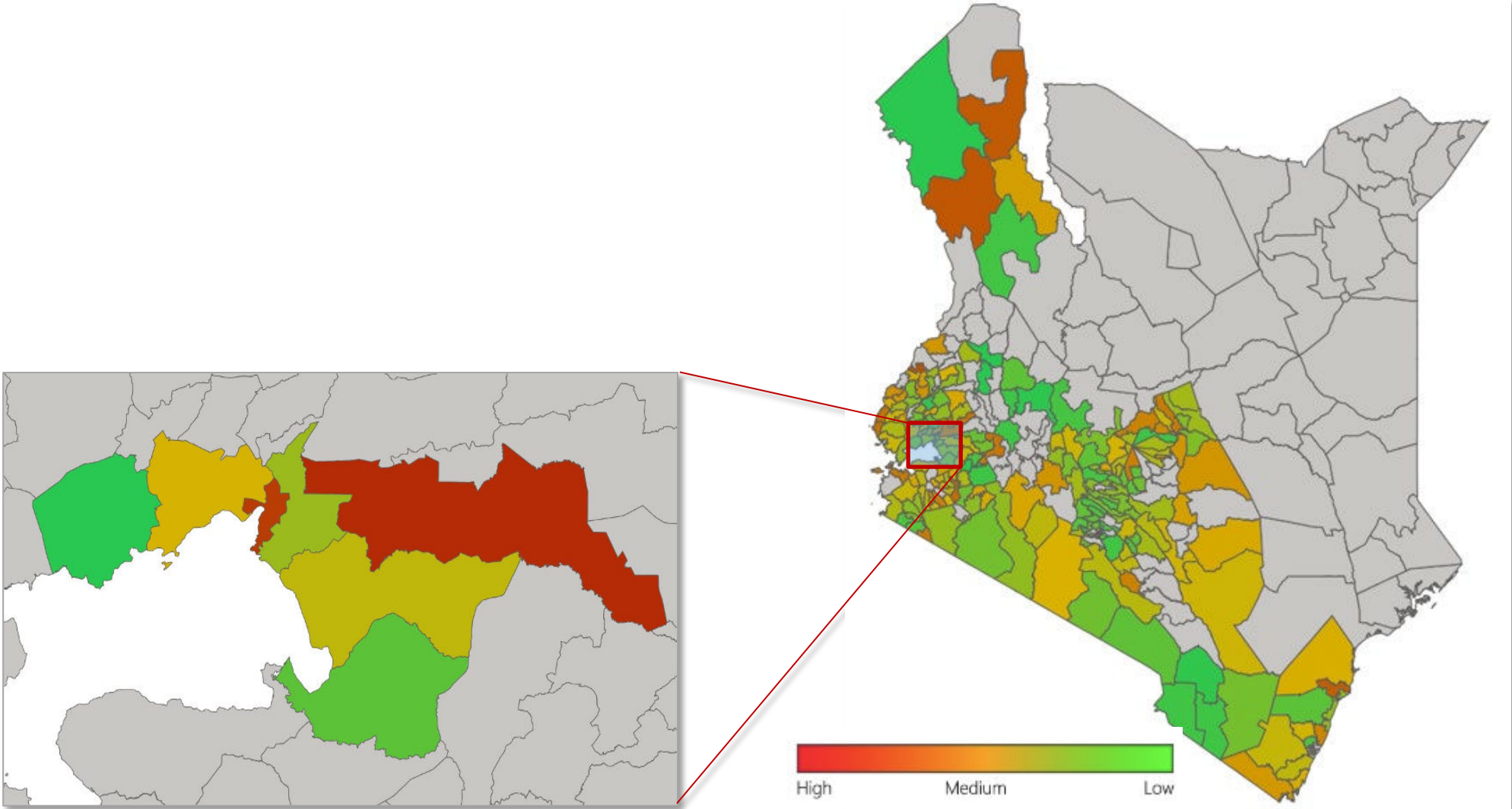
How does it work?

- HTS providers generate the probability of positive results and are notified of HIV testing recommendation based on client's risk.
- Providers can refer clients for HIV prevention for high-risk clients that turn HIV negative.

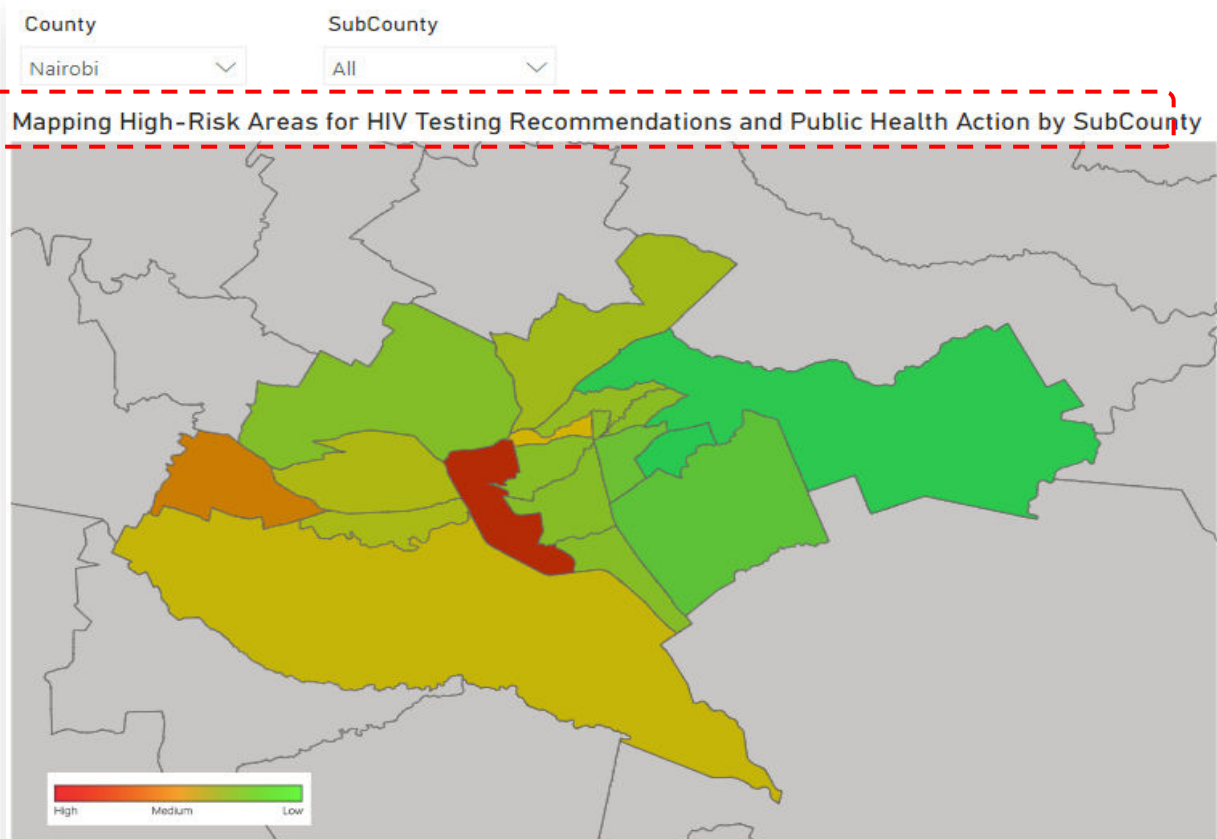
Use Case 1: Use of ML to Track Linkage to Prevention Services by Risk Profile

Referral Service	High or Very High				Low or Moderate			
	Female (N = 131,152)		Male (N = 60,969)		Female (N = 613,433)		Male (N = 292,996)	
Comprehensive care center	2,917	2.22%	1,704	2.79%	3,042	0.50%	1,856	0.63%
Condom use counselling	33,376	25.45%	18,739	30.74%	129,778	21.16%	73,723	25.16%
Confirmatory test	3,260	2.49%	1,877	3.08%	3,549	0.58%	2,083	0.71%
DBS for PCR	10	0.01%	1	0.002%	9	0.001%	7	0.002%
HIV testing and re-testing	13,011	9.92%	6,914	11.34%	77,944	12.71%	29,077	9.92%
Post-exposure prophylaxis	6,233	4.75%	3,606	5.91%	21,794	3.55%	14,292	4.88%
Pre-Exposure Prophylaxis	19,669	15.00%	10,468	17.17%	63,380	10.33%	34,120	11.65%
Prevention and treatment of STIs	8,408	6.41%	4,966	8.15%	44,277	7.22%	19,364	6.61%
Prevention of GBV	6,918	5.27%	3,158	5.18%	41,815	6.82%	13,452	4.59%
Risk reduction counselling	48,932	37.31%	24,437	40.08%	228,561	37.26%	113,476	38.73%
Safer sex practices	38,511	29.36%	19,431	31.87%	180,601	29.44%	87,553	29.88%
Substance abuse and mental health treatment	4,246	3.24%	2,779	4.56%	18,287	2.98%	10,924	3.73%
VMMC	-	-	3,059	5.02%	-	-	14,064	4.80%
Other	4,392	3.35%	1,850	3.03%	22,084	3.60%	7,689	2.62%

Use Case 2: Use of ML for Geographic Clustering of Risk Profiles



Use Case 3: Use of ML for Geographic and Demographic Clustering of Risk Profile



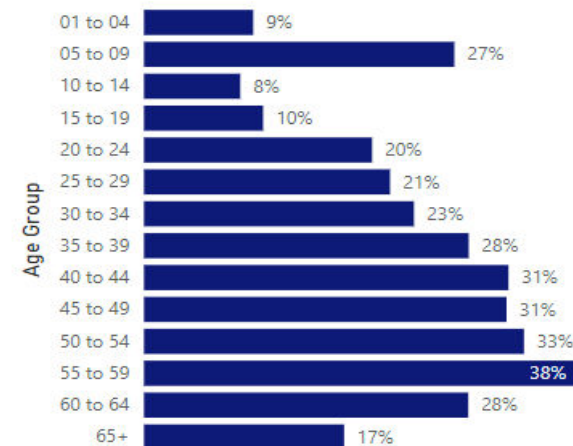
Description: This dashboard provides a visual representation of areas color-coded based on the percentage of high-risk individuals for HIV testing. Powered by machine learning, it offers targeted recommendations for testing in high-risk sub counties, facilitating strategic public health interventions to prevent HIV.

% of High Risk Category Among HIV Tests by Gender



Distribution of Risk by Sex

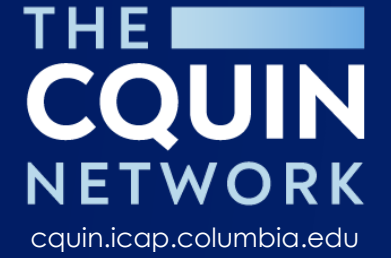
% of High Risk Category Among HIV Tests by Age Group



Distribution of Risk by Age

Summary

- Useful tools to tailor services and interventions
- Investment in data systems is key:
 - Data collection
 - Data transmission
 - Data storage
 - Data Privacy
- Multidisciplinary team engagement
- Program- Strategic direction, end user feedback on the look and feel of systems
- HIS-System architecture, model development, testing, deployment and monitoring



Thank You!

